

# Multifractal stationary random processes

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# Multifractal stationary random processes

- Random cascades (Mandelbrot)
- Intuitive construction
- Theoretical construction
- Some particular examples

# Random cascades (Mandelbrot)

Dyadic case

- $l_n = T\lambda^n$ ,  $\lambda = 1/2$
- $I = [0, T[, I_0 = [0, T/2[, I_{01} = [T/4, T/2[, \dots$
- $M_{l_n}(I_{s_0 \dots s_n}) = W_{s_0 \dots s_n} M_{l_{n-1}}(I_{s_0 \dots s_{n-1}})$
- $\{W_{s_0 \dots s_i}\}$  are i.i.d,
- $\mathbb{E}(W_{s_0 \dots s_i}) = 1/2$  ( $\Rightarrow M_{l_n}$  martingale).
- “Discrete scale invariance” (around 0)

$$\int_0^{t/2} M_{l_n}(dt) \stackrel{law}{=} W \int_0^t M_{l_{n-1}}(dt)$$

$$M_{l_n}(dt/2) \stackrel{law}{=} W M_{l_{n-1}}(dt)$$

$$M_{l_n}(dt l_n / l_{n-1}) \stackrel{law}{=} W M_{l_{n-1}}(dt)$$

By taking  $n \rightarrow +\infty$

$$\int_0^{t/2} M(dt) \stackrel{law}{=} W \int_0^t M(dt)$$

→ NO : not stationary, no continuous scale invariance

# Towards “continuous” cascades

Discrete cascades (dyadic case)

$$M_{l_n}(d t l_n / l_{n-1}) \stackrel{law}{=} W M_{l_{n-1}}(d t)$$

$\implies$  Try to represent it as a discretization of an underlying continuous construction.

- $M_{l'}(l' d t) \stackrel{law}{=} W_{l',l}(t) M_l(l d t), \quad l' \leq l \leq T$
- law of  $W_{l',l}(t)$  depends only on  $l'/l$  (and *i.i.d*).

# Why infinitely divisible laws?

- $M_l(dt) \stackrel{law}{=} W_{l,l'}(t)M_{l'}(dt), \quad l \leq l' \leq T$

$$\Rightarrow \ln W_{l',T} = \ln W_{l',l} + \ln W_{l,T}$$

$$\Rightarrow \ln W_{l',T} = \text{sum of 2 i.i.d variables (for } l = \sqrt{l'T} \text{)}$$

$$\omega_l(t) = \ln W_{l,T}(t) = \sum \text{arbitrary number of i.i.d var.}$$

$\Rightarrow \omega_l(t)$  is infinitely divisible.

- $P(dt, dl)$  stochastic “infinitely divisible noise” on the  $\mathcal{S}^+$  half-plane  $(t, l) \in \mathbb{R} \times \mathbb{R}^{+*}$  with respect to the measure  $\mu(dt, dl)$ ,

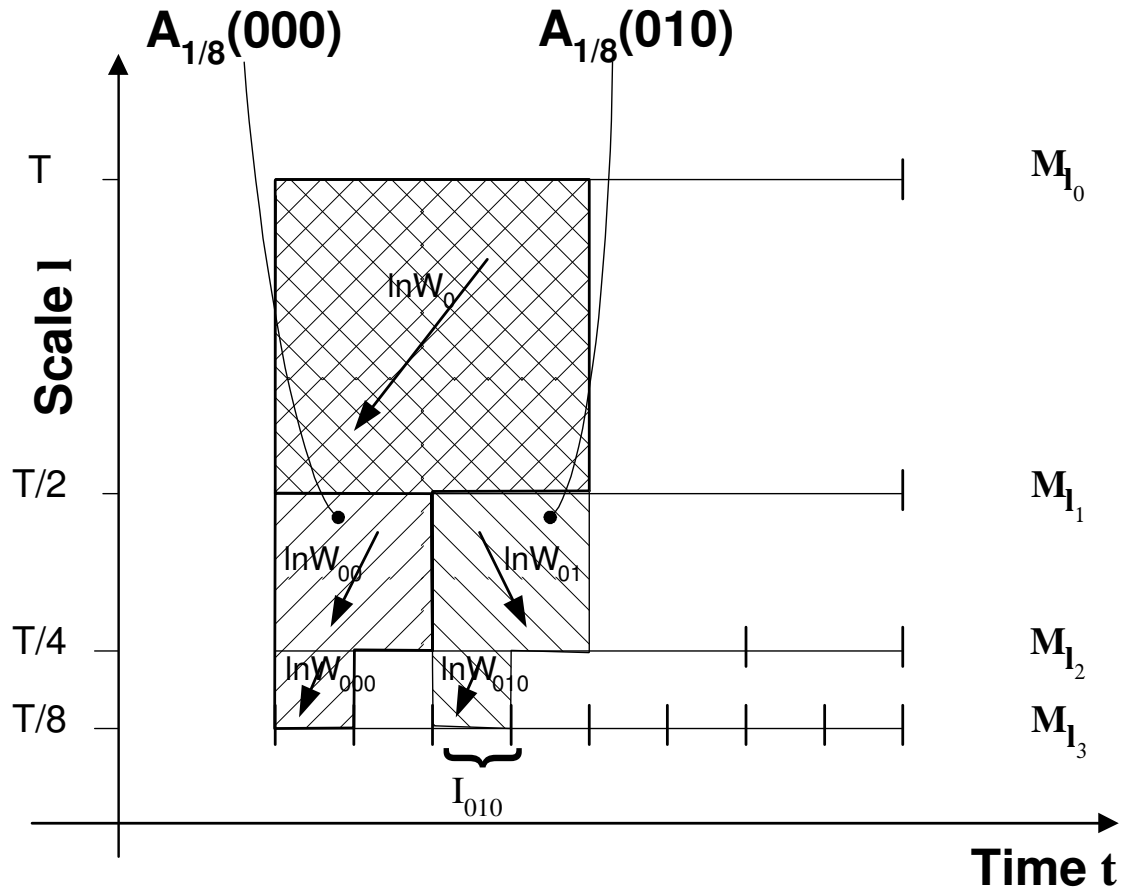
- $A \subset \mathcal{S}^+, P(A)$  infinitely divisible

- the law of  $P(A)$  depends only on  $\mu(A)$

- $P(A)$  and  $P(B)$  independent iff  $A \cap B = \emptyset$

# 2d-representation of discrete cascades

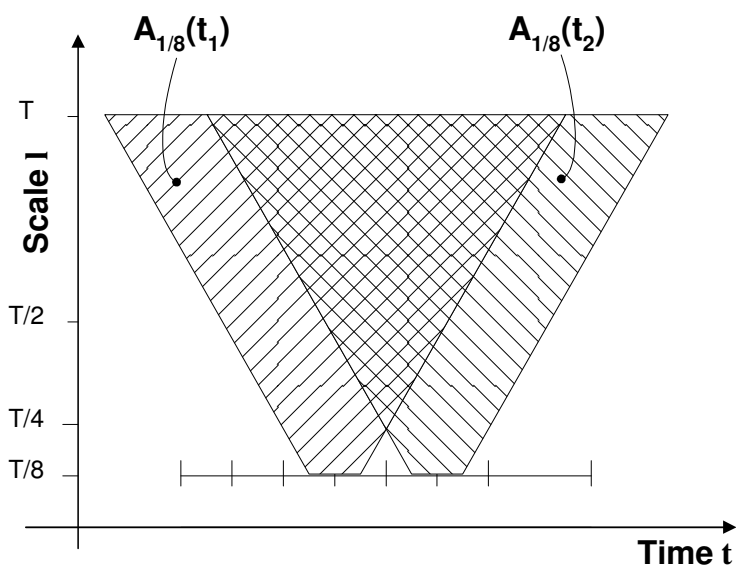
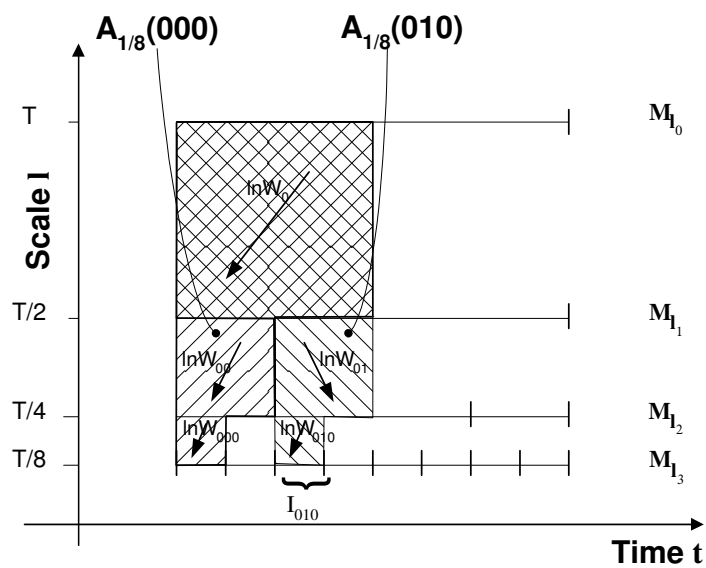
$$\omega_{l_n}(s_0 \dots s_n) = \sum_{i=0}^n \ln(W_{s_0 \dots s_i})$$



$$\omega_{l_n}(s_0 \dots s_n) = P(A_{2^{-n}}(s_0 \dots s_n))$$

$$\ln W_{s_0 \dots s_i} \text{ i.d.} \Rightarrow \mu(dt, dl) = dt dl / l^2$$

# Continuous cascades



$$\omega_l(t) = \ln W_{l,T}(t) = P(A_l(t))$$

$$M_l(dt) \stackrel{def}{=} e^{\omega_l(t)} dt$$

$$\implies M_{l'}(dt) =^{law} W_{l',l}(t) M_l(dt), \quad l' \leq l \leq T$$



# Infinitely divisible laws

- **Definition**

*$X$  is infinitely divisible iff it is a sum of an arbitrary numbers of i.i.d. variables*

- **Examples**

- Gaussian variable (mean  $m$  , variance  $\sigma^2$  )

$$p(x) = e^{-\frac{(x-m)^2}{\sigma^2}} / \sqrt{2\pi\sigma^2}$$

$$\mathbb{E}(e^{iqX}) = e^{imq - q^2\sigma^2/2}$$

$$\phi_X(q) = \ln \mathbb{E}(e^{iqX}) = imq - q^2\sigma^2/2$$

$\phi_X(q)$  is the *cumulant generating function* of  $X$

- Poisson variable (intensity  $\lambda$  )

$$p(x) = e^{-\lambda} \lambda^k / k!$$

$$\phi_X(q) = \lambda(e^{iq} - 1)$$

- **Theorem**

*A sum of independent infinitely divisible variables is an infinitely divisible variable.*

*Moreover  $\phi_{\alpha X + \beta Y}(q) = \phi_X(\alpha q) + \phi_Y(\beta q)$ .*

# The Levy-Khintchine formula

- Combining gaussian and (an infinite number of) poisson variables :

$$\phi(q) = imq - q^2\sigma^2/2 + \int (e^{iqx} - 1)\tilde{\nu}(dx)$$

$$\tilde{\nu}(dx) \simeq \text{“intensity measure”}$$

$$\text{a priori, total intensity bounded: } \int \tilde{\nu}(dx) < +\infty$$

$$\text{actually, intensity around 0 can be } +\infty \quad (x^{-2}dx)$$

- Levy-Khintchine representation

$$\phi(q) = imq + \int \frac{e^{iqx} - 1 - iq \sin(x)}{x^2} \nu(dx)$$

→  $\nu(dx)$  is the *Levy measure*

$$\rightarrow \int_{-y}^y \nu(dx), \int_{-\infty}^{-y} \nu(dx)/x^2, \int_y^{+\infty} \nu(dx)/x^2 < +\infty$$

–  $\nu(dx) = \delta(x)$  → gaussian

–  $\nu(dx) = \delta(x - x_0)$  → poisson

–  $\int \nu(dx)/x^2 < +\infty$  → compound poisson

–  $\alpha$ -stable, gamma, Student, ...

# “Infinitely divisible noise”

**Definition** (*Rosinski, 1989*)

$P(dt, dl)$  is an *independently scattered infinitely divisible random measure* distributed on the half plane  $\mathcal{S}^+$  with respect to the (deterministic measure)  $\mu(dt, dl)$

- $\{\mathcal{A}_n\}_n$  disjoint sets of  $\mathcal{S}^+$ ,  $\{P(\mathcal{A}_n)\}_n$  are i.i.d.

$$P\left(\bigcup_{n=1}^{\infty} \mathcal{A}_n\right) = \sum_{n=1}^{\infty} P(\mathcal{A}_n), \quad \text{a.s.}$$

- $P(\mathcal{A})$  is infinitely divisible with

$$\mathbb{E}\left(e^{iqP(\mathcal{A})}\right) = e^{\phi(q)\mu(\mathcal{A})},$$

**Notation :**  $\psi(q) = \phi(-iq)$  whenever it is possible and  $\psi(q) = +\infty$  otherwise ( $\psi(q)$  is convex)

$$\rightarrow \psi(q) = +\infty, \quad \text{if } \mathbb{E}\left(e^{qP(\mathcal{A})}\right) = +\infty,$$

$$\rightarrow \mathbb{E}\left(e^{qP(\mathcal{A})}\right) = e^{\psi(q)\mu(\mathcal{A})}, \quad \text{otherwise.}$$

# Continuous cascades

$$M_l(dt) \stackrel{\text{def}}{=} e^{P(A_l(t))} dt$$

- *Schmitt F. and Marsan D., 2001*
  - basic ideas for construction of  $\omega_l(t) = P(\mathcal{A}_l(t))$
  - scaling of  $e^{\omega_l}$  versus  $l$
- *Bacry E., Delour J. and Muzy J.F., 2001*
  - 1d representation of the log-normal case (MRW)
  - exact power law scaling of the moments
  - no proof of convergence
- *Barral J. and Mandelbrot B.B., 2002*
  - full 2d compound poisson construction
  - non degeneracy,  $L^p$  convergence, ...
  - multifractal formalism
- *Bacry E., Muzy J.F., 2003*
  - full 2d log-inf. divisible construction (MRM, MRW)
  - non degeneracy,  $L^p$  convergence, ...
  - exact/asymptotic scaling of the moments
- *Abry P. , Chainais P., Riedi R. 2003*
  - full 2d log-infinitely divisible construction (IDC)
  - $L^2$  convergence
  - non power law scaling of the moments

# Multifractal Random Measures (MRM)

(Bacry, Muzy, 2003)

- $\mathcal{A}_l(t)$  domain :

- General definition

$$\mathcal{A}_l(t) = \{(t', l'), l' \geq l, |t' - t| \leq f(l')/2\}.$$

$$\int_l^\infty f(s)/s^2 ds < \infty \quad \text{and} \quad f(l) = l \quad \text{for } l < L$$

- Definition for “exact scaling”

$$f^{(e)}(l) = l \quad \text{for } l \leq T \quad \text{and} \quad f^{(e)}(l) = T \quad \text{for } l \geq T$$

- $\omega_l(t) = P(\mathcal{A}_l(t))$

- Levy measure  $\nu(dx)$

- $\psi(1) = 0$  ( $e^{\omega_l(t)}$  martingale)

- $\exists \epsilon > 0, \psi(1 + \epsilon) < +\infty$  ( $\mathbb{E}(e^{(1+\epsilon)\omega_l(t)}) < +\infty$ )

- $M_l(dt) = e^{\omega_l(t)} dt$

- $M(dt) = \lim_{l \rightarrow 0+} M_l(dt)$

# Existence of the limit MRM measure

*(Bacry, Muzy, 2003)*

**Theorem 1** *There exists a measure  $M(dt)$  such that*

*(i) with probability one,  $M_l(dt) \rightarrow M(dt)$  weakly*

*(ii)  $\forall t \in \mathbb{R}, M(\{t\}) = 0,$*

*(iii) for any bounded set  $K$  of  $\mathbb{R},$*

*-  $M(K) < +\infty$*

*-  $\mathbb{E}(M(K)) \leq |K|.$*

*Proof*

- $\psi(1) = 0 \implies \mathbb{E}(e^{\omega_l(t)}) = 1.$
- $\{M_l(I)\}_l$  is a left continuous positive martingale
- Kahane J.P., *Chi. Ann. of Math.* **8B**, 1-12 (1987)

□

# Continuous scale invariance

(Bacry, Muzy, 2003)

$$M_{l'}(l' dt) = W_{l',l}(t) M_l(l dt), \quad l' \leq l \leq T.$$

$$\implies M_{\lambda l}(\lambda l dt) = W_{\lambda}(t) M_l(l dt), \quad l' = \lambda l,$$

$$\implies M_{\lambda l}(\lambda dt) = W_{\lambda}(t) M_l(dt),$$

$$\implies M_{\lambda l}([0, \lambda t]) = W_{\lambda}(0) M_l([0, t]).$$

On the one hand

$$M_{\lambda l}([0, \lambda t]) = \int_0^{\lambda t} e^{\omega_{\lambda l}(u)} du = \lambda \int_0^t e^{\omega_{\lambda l}(\lambda u)} du.$$

On the other hand

$$W_{\lambda}(0) M_l([0, t]) = W_{\lambda}(0) \int_0^t e^{\omega_l(u)} du.$$

Consequently we want

$$\lambda e^{\omega_{\lambda l}(\lambda t)} \stackrel{law}{=} W_{\lambda} \int_0^t e^{\omega_l(u)} du$$

Or,

$$\omega_{\lambda l}(\lambda t) \stackrel{law}{=} \ln(W_{\lambda}/\lambda) + \omega_l(t)???$$

# Scaling of $\omega_l(t)$

(Bacry, Muzy, 2003)

We want  $\{\omega_{\lambda l}(\lambda t)\}_{t \leq T} \stackrel{law}{=} \ln(W_\lambda/\lambda) + \{\omega_l(t)\}_{t \leq T}$ ,

**Lemma 1 (Characteristic function of  $\omega_l(t)$ )**

$$\mathbb{E} \left( e^{\sum_{m=1}^q i p_m \omega_l(t_m)} \right) = e^{\sum_{j=1}^q \sum_{k=1}^j \alpha(j,k) \rho_l(t_k - t_j)},$$

where

- $\rho_l(t) = \mu(\mathcal{A}_l(0) \cap \mathcal{A}_l(t).)$ ,
- $\sum_{j=1}^q \sum_{k=1}^j \alpha(j,k) = \varphi \left( \sum_{k=1}^q p_k \right).$

□

The exact scaling domain  $f^{(e)}(l)$  is the only domain which satisfies

$$\rho_{\lambda l}(\lambda t) = -\log \lambda + \rho_l(t), \quad l \leq T, \quad \lambda < 1, \quad t < T$$

In this case,  $\ln(W_\lambda/\lambda)$  is infinitely divisible (indep. of  $M$ ) with

$$\mathbb{E} \left( e^{iq \ln(W_\lambda/\lambda)} \right) = \lambda^{-\varphi(q)}.$$



# Continuous scale invariance of $M^{(e)}(t)$

(Bacry, Muzy, 2003)

## Theorem 2 (Continuous invariance of $M^{(e)}(t)$ )

$$\{M^{(e)}([0, \lambda t])\}_{t \leq T} \stackrel{law}{=} W_\lambda \{M^{(e)}([0, t])\}_{t \leq T},$$

where  $\ln(W_\lambda/\lambda)$  is infinitely divisible (indep. of  $M$ ) with

$$\mathbb{E} \left( e^{iq \ln(W_\lambda/\lambda)} \right) = \lambda^{-\varphi(q)}.$$

## Theorem 3 (Moment scaling of $M^{(e)}(t)$ )

$$\mathbb{E} \left( M^{(e)}([0, t])^q \right) = \left( \frac{t}{T} \right)^{\zeta(q)} \mathbb{E} \left( M^{(e)}([0, T])^q \right), \quad \forall t \leq T.$$

where

$$\zeta(q) = q - \psi(q)$$

# Moments, Degeneracy of $M^{(e)}(t)$

(Bacry, Muzy, 2003)

## Theorem 4 (Existence of the moments of $M^{(e)}(t)$ )

Let  $q > 0$  then

(i)  $\zeta(q) > 1 \implies \mathbb{E} \left( M^{(e)}([0, t])^q \right) < +\infty$  and

$$\sup_t \mathbb{E} \left( M_t^{(e)}([0, t])^q \right) < +\infty.$$

(ii) if  $M^{(e)} \neq 0$ ,  $\mathbb{E} \left( M^{(e)}([0, t])^q \right) < +\infty \implies \zeta(q) \geq 1$ .

*Proof for (ii)*

$$\begin{aligned} \mathbb{E} \left( M^{(e)}([0, t])^q \right) &= \mathbb{E} \left( (M^{(e)}([0, t/2]) + M^{(e)}([t/2, t]))^q \right) \\ &\geq \mathbb{E} \left( M^{(e)}([0, t/2])^q \right) + \mathbb{E} \left( M^{(e)}([t/2, t])^q \right) \end{aligned}$$

Using scaling of the moments

$$\mathbb{E} \left( M^{(e)}([0, t])^q \right) \geq 2^{1-\zeta(q)} \mathbb{E} \left( M^{(e)}([0, t])^q \right),$$

and consequently  $\zeta(q) \geq 1$ .

□

## Theorem 5 (Non degeneracy of $M^{(e)}(t)$ )

(H)  $\exists \epsilon > 0, \zeta(1 + \epsilon) > 1$

if (H) holds then  $\mathbb{E} \left( M^{(e)}([0, t]) \right) = t$ .

# Summing up results on $M^{(e)}(t)$

*(Bacry, Muzy, 2003)*

- (i)  $\psi(1) = 0 \implies$  existence of  $M^{(e)}$
- (ii)  $\exists \epsilon > 0, \zeta(1 + \epsilon) > 1 \implies \mathbb{E} (M^{(e)}([0, t])) = t$
- (iii)  $\zeta(q) > 1$  “ $\iff$ ”  $\mathbb{E} (M^{(e)}([0, t])) < +\infty$
- (iv)  $\{M^{(e)}([0, \lambda t])\}_{t \leq T} =^{law} W_\lambda \{M^{(e)}([0, t])\}_{t \leq T},$
- (v)  $\mathbb{E} (M^{(e)}([0, t])^q) = \left(\frac{t}{T}\right)^{\zeta(q)} \mathbb{E} (M^{(e)}[0, T]^q), \quad \forall t \leq T.$

# From $M^{(e)}(t)$ to $M(t)$

(Bacry, Muzy, 2003)

**Theorem 6** (Degeneracy, asymptotic scaling and moments of positive orders of  $M(dt)$ )

(i)  $M^{(e)}(dt) \stackrel{a.s.}{=} 0 \iff M(dt) \stackrel{a.s.}{=} 0,$

Moreover, if  $M(dt) \neq 0$ , one has

(ii) with probability one,

$$\forall t \geq 0, \quad M([0, \lambda t]) \sim X M^{(e)}([0, \lambda t]) \text{ , when } \lambda \rightarrow 0^+.$$

(iii)  $\mathbb{E} \left( M^{(e)}([0, t])^q \right) < +\infty \iff \mathbb{E} \left( M([0, t])^q \right) < +\infty.$

(iv)  $\mathbb{E} \left( M([0, t])^q \right) \sim_{t \rightarrow 0^+} \left( \frac{t}{T} \right)^{\zeta(q)} \mathbb{E} \left( M([0, T])^q \right).$

□

# Multifractal Random Processes (MRW)

*(Bacry, Muzy, 2003)*

- **Definition (subordinated process)**

$$X^{(s)}(t) = B(M([0, t])).$$

- **Equivalent definition (stochastic integral)**

$$X(t) = \lim_{l \rightarrow 0^+} X_l(t),$$

where

$$X_l(t) = \int_0^t e^{\omega_l(u)/2} dW(u),$$

# Main theorem on MRW

(Bacry, Muzy, 2003)

**Theorem 7** *Under hypothesis  $\zeta(1 + \epsilon) > 1$  (non degeneracy of  $M$ ),*

$$(i) \quad \zeta(q) > 1 \implies \mathbb{E}(|X(t)|^{2q}) < +\infty.$$

$$(ii) \quad \mathbb{E}(|X(t)|^{2q}) < +\infty \implies \zeta(q) \geq 1.$$

$$(iii) \quad \{X^{(e)}(t)\}_{t \leq T} =^{law} W_\lambda \{X^{(e)}(t)\}_{t \leq T},$$

$$(iv) \quad \mathbb{E}(|X^{(e)}(t)|^{2q}) = \left(\frac{t}{T}\right)^{\zeta(q)} \mathbb{E}(|X^{(e)}(T)|^{2q}), \quad \forall t \leq T.$$

$$(v) \quad \mathbb{E}(|X(t)|^{2q}) \sim_{t \rightarrow 0^+} \left(\frac{t}{T}\right)^{\zeta(q)} \mathbb{E}(|X(T)|^{2q}).$$

# Numerical simulation of MRW

(Bacry, Muzy, 2003)

- $\{\epsilon[k]\}_{k \in \mathbb{Z}}$  Gaussian white noise, variance 1.
- $l_n = 2^{-n}$ .
- $\tilde{X}_{l_n}(t) = \sum_{k=0}^{t/l_n} \sigma \sqrt{l_n} e^{\frac{\omega_{l_n}(kl_n)}{2}} \epsilon[k]$ .

**Theorem 8 (Convergence)** *If  $\zeta(2 + \epsilon) > 1$  then*

$$\lim_{n \rightarrow +\infty} \{\tilde{X}_{l_n}(t)\}_t \stackrel{law}{=} \{X(t)\}_t.$$

*Proof*

$$\lim_{n \rightarrow +\infty} \tilde{M}_{l_n}(dt) =^{m.s.} M(dt)$$

$$\text{where } \tilde{M}_{l_n}([0, t]) = \sum_{k=0}^{t/l_n} e^{\omega_{l_n}(kl_n)} l_n$$

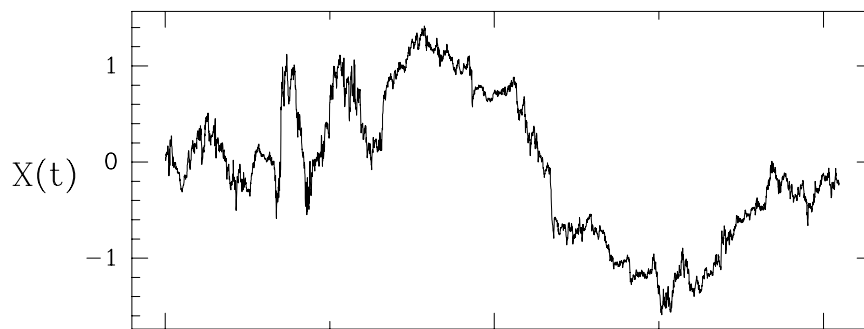
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$\implies X(t)$  is the continuous limit of a random walk with stochastic variance

# Log-normal MRW

(Bacry, Muzy, 2003)

- $\nu(dx) = 2\lambda^2\delta(x),$
- $\psi(q) = qm + \lambda^2q^2$  with  $m = -\lambda^2.$
- $\zeta(q) = q(1 + \lambda^2) - \lambda^2q^2.$



$\sigma^2 = 1, T = 512$  and  $\lambda^2 = 0.025.$

- **1d construction :** *Bacry, Delour, Muzy, 2001*

$$X_l(t) = X_l(t - l) + \sigma\sqrt{l}e^{\frac{\omega_l(t)}{2}}\epsilon_l(t), \quad (\text{step-wise process})$$

$$\mathbb{E}(\omega_l) = -\text{Var}(\omega_l)/2$$

$$\text{Cov}(\omega_l(0), \omega_l(t)) = \begin{cases} 4\lambda^2 \ln\left(\frac{T}{(|t|+l)}\right) & \text{for } |t| \leq T \\ 0 & \text{otherwise} \end{cases}$$

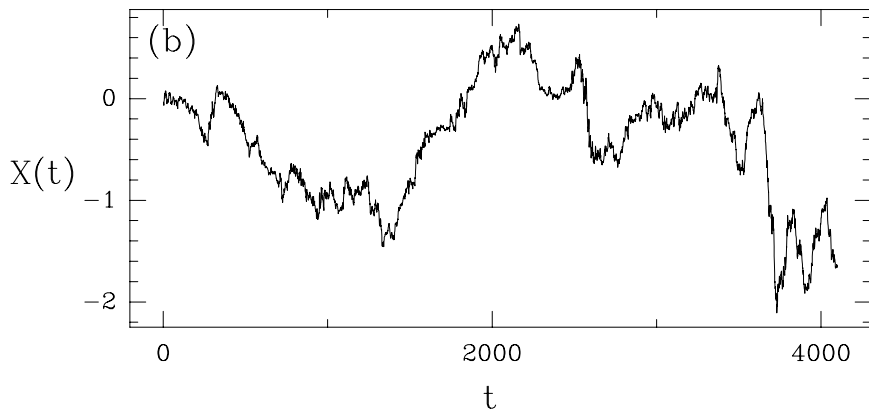
$$X(t) = \lim_{l \rightarrow 0^+} X_l(t)$$



# Log-Poisson MRW

(Bacry, Muzy, 2003 and  $\simeq$  Barral, Mandelbrot, 2002)

- $\nu(dx) = \gamma \delta(x - x_0)$ , with  $\gamma = \lambda x_0^2$ ,  $x_0 = \ln \delta$
- $\zeta(q) = qm + \lambda(1 - \delta^q)$  ( $m$  such that  $\zeta(1) = 1$ )



$\sigma^2 = 1$ ,  $T = 512$ ,  $\lambda = 4$  and  $x_0 = 0.05$ .

- Log-Poisson compound

$$\int \nu(dx) x^{-2} = C < +\infty, p_{\ln W} = \nu(dx) x^{-2} / C$$

Compounded with  $\ln W$ , intensity  $C$

$$\zeta(q) = qm + C(1 - \mathbb{E}(W^q)), \quad (m \text{ such that } \zeta(1) = 1)$$

# Other MRW's

(Bacry, Muzy, 2003)

- log- $\alpha$ -stable MRW

left-sided  $\alpha$ -stable density :

$$\nu(dx) = \begin{cases} C|x|^{1-\alpha} & \text{if } x \leq 0 \\ 0 & \text{if } x > 0 \end{cases}$$

with  $C > 0$  and  $0 < \alpha < 2$

$$\zeta(q) = qm - \sigma^\alpha |q|^\alpha$$

stable  $\Rightarrow$  1d construction

- log-gamma MRW

$$\nu(dx) = C\gamma^2 x e^{-\gamma x} dx \text{ for } x \geq 0$$

$$\zeta(q) = qm - C\gamma^2 \ln \frac{\gamma}{\gamma-q}, \text{ with } \gamma > 1$$

- log-Student, log-Pareto, ...

# Log-normal MRW

(Bacry, Muzy, 2001, 2003)

- Continuous limit of a rand. walk with stoch. variance

$$X_l(t) = X_l(t-l) + \sigma^2 \sqrt{l} e^{\frac{\omega_l(t)}{2}} \epsilon_l(t), \quad (\text{step-wise})$$

$$\mathbb{E}(\omega_l) = -\text{Var}(\omega_l)/2$$

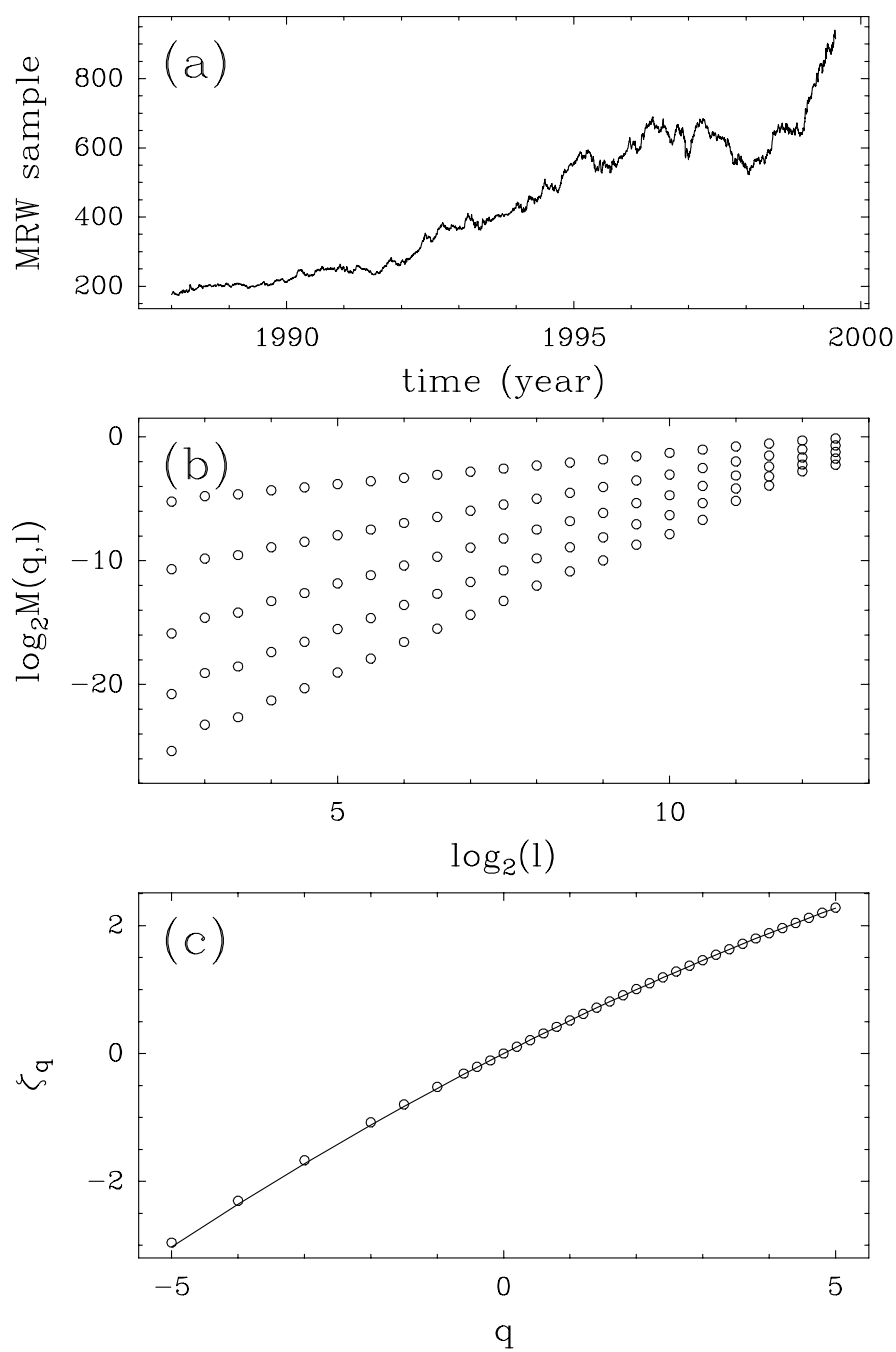
$$\text{Cov}(\omega_l(0), \omega_l(t)) = \begin{cases} 4\lambda^2 \ln\left(\frac{T}{(|t|+l)}\right) & \text{for } |t| \leq T \\ 0 & \text{otherwise} \end{cases}$$

$$X(t) = \lim_{l \rightarrow 0^+} X_l(t)$$

- $\zeta(q) = q - q(q-1)\lambda^2$
- Only 3 parameters :
  - $\sigma^2$  : deterministic random walk variance
  - $T$  : correlation length
  - $\lambda^2$  : “multifractal coefficient”  
 $\lambda = 0 \Rightarrow X(t) = B(t)$

# Multifractal analysis of a MRW

*(Bacry, Delour, Muzy, 2001)*



# Moments and law deformation across scales

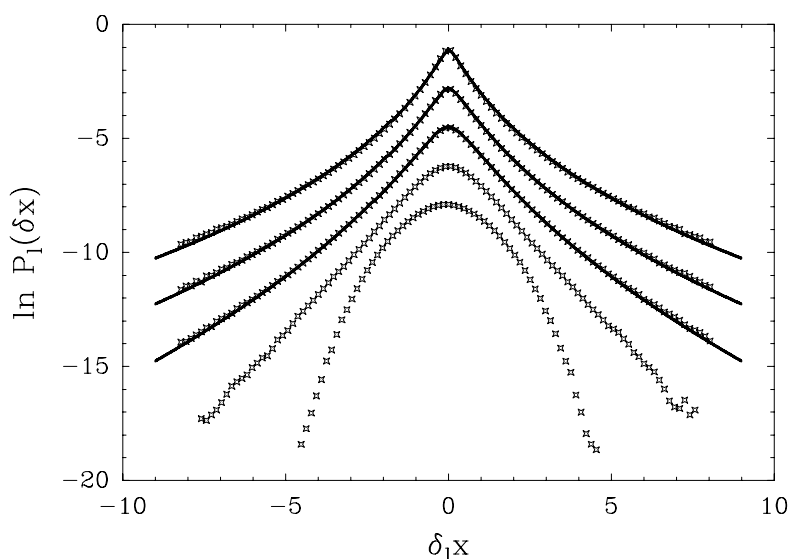
(Bacry, Muzy, 2001, 2003)

- Explicit  $q$ -order moment formula

$$\mathbb{E} (X(t)^{2q}) = C_{2q} T^q \sigma^{2q} (2q-1)!! \left( \frac{t}{T} \right)^{\zeta(q)}, \quad \text{for } t \leq T$$

$$C_q = \prod_{k=0}^{q/2-1} \frac{\Gamma(1-2\lambda^2 k)^2 \Gamma(1-2\lambda^2(k+1))}{\Gamma(2-2\lambda^2(q/2+k-1)) \Gamma(1-2\lambda^2)}$$

- $q > 1/\lambda^2$  "  $\iff$  "  $\mathbb{E} (X(t)^{2q}) = +\infty$



MRW

oooo Numerical estimation

— log-normal approx.

# Discretization

(Bacry, Kozhemyak, Muzy, 2004)

- Modeling discrete data with stationary increments
  - Discrete model  $X_\Delta[n] = X(n\Delta)$
  - Stationary incr.  $\delta X_\Delta[n] = X_\Delta[n+1] - X_\Delta[n]$

- finance, turbulence, ...:

$$\text{Curvature}(\zeta(q)) = \lambda^2 \ll 1$$

- If  $\lambda^2 \ll 1$ ,  $X_\Delta[n]$  “is close to” rand. walk  $R_\Delta[n]$

$$\delta R_\Delta[n] = \epsilon[n] e^{\lambda \Omega_\Delta[n]/2},$$

$\epsilon[n]$  : white noise,  $\Omega_\Delta[n]$  : known Gaussian process.

## Theorem 9 (Meaning of “is close to”)

$$\left\{ \frac{2 \ln M([(n-1)\Delta, n\Delta])}{\lambda} \right\}_n \xrightarrow[\lambda \rightarrow 0^+]{law} \{\Omega_\Delta[n]\}_n$$

Moreover,

- $\mathbb{E}(\ln |\delta X_\Delta[n]|^q \dots) = \mathbb{E}(\ln |\delta R_\Delta[n]|^q \dots) (1 + o(\lambda^{2-\epsilon}))$

and when  $\mathbb{E}(\delta X_\Delta[n]^q \dots) < +\infty$

- $\mathbb{E}(\delta X_\Delta[n]^q \dots) = \mathbb{E}(\delta R_\Delta[n]^q \dots) (1 + o(\lambda^{2-\epsilon})),$

# Application to Parameter Estimation

(Bacry, Kozhemyak, Muzy, 2004)

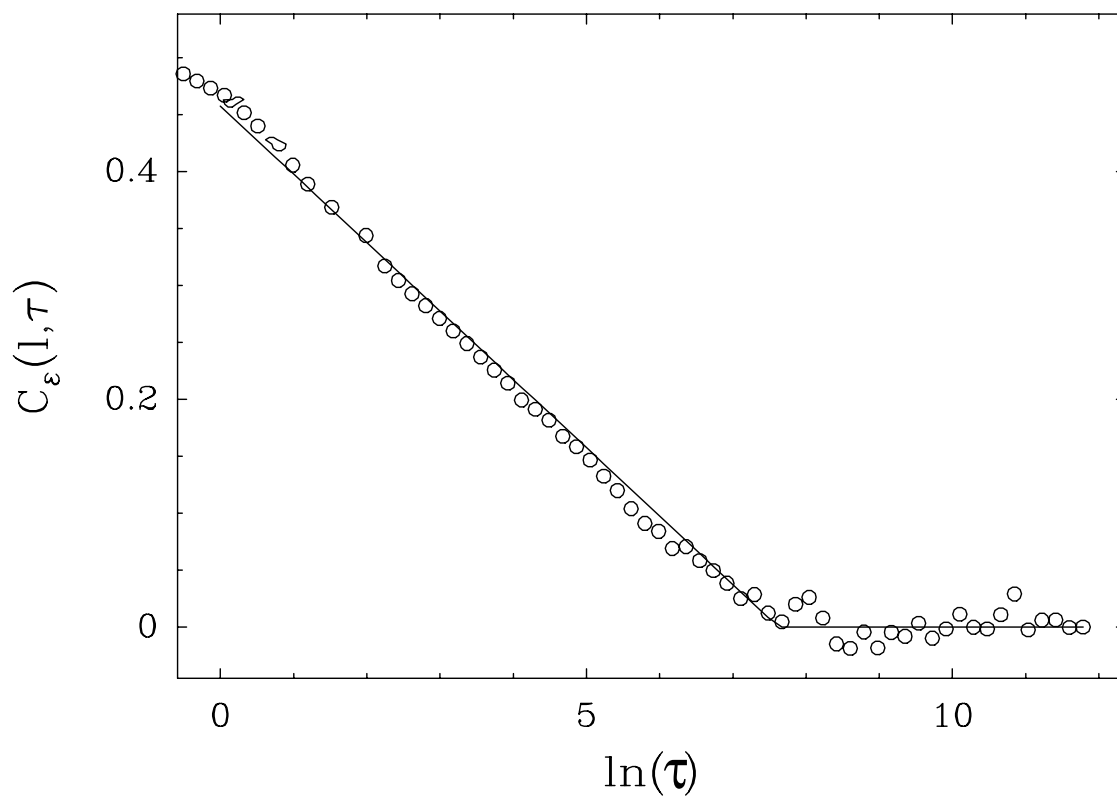
$$\begin{aligned} & Cov(\ln |\delta X_\Delta[0]|, \ln |\delta X_\Delta[\tau]|) \\ & \simeq Cov(\ln |\delta R_\Delta[0]|, \ln |\delta R_\Delta[\tau]|) \\ & \simeq Var(\ln |\epsilon|) + \lambda^2 Cov(\Omega_\Delta[0], \Omega_\Delta[\tau])/4 \\ & \simeq -\lambda^2 \ln(\tau/T), \quad \text{for } 1 < \tau < T \\ & \implies \text{independent of } \lambda \end{aligned}$$

- Estimation of the multifractal coefficient  $\lambda^2$   
→ linear fit
- Integral scale  $T$   
→ decorrelation scale
- Variance  $\sigma^2$   
→ variance of the increments

$\implies$  GMM estimation

# Correlation function of $\ln |\delta X[n]|$

*(Bacry, Kozhemyak, Muzy, 2004)*

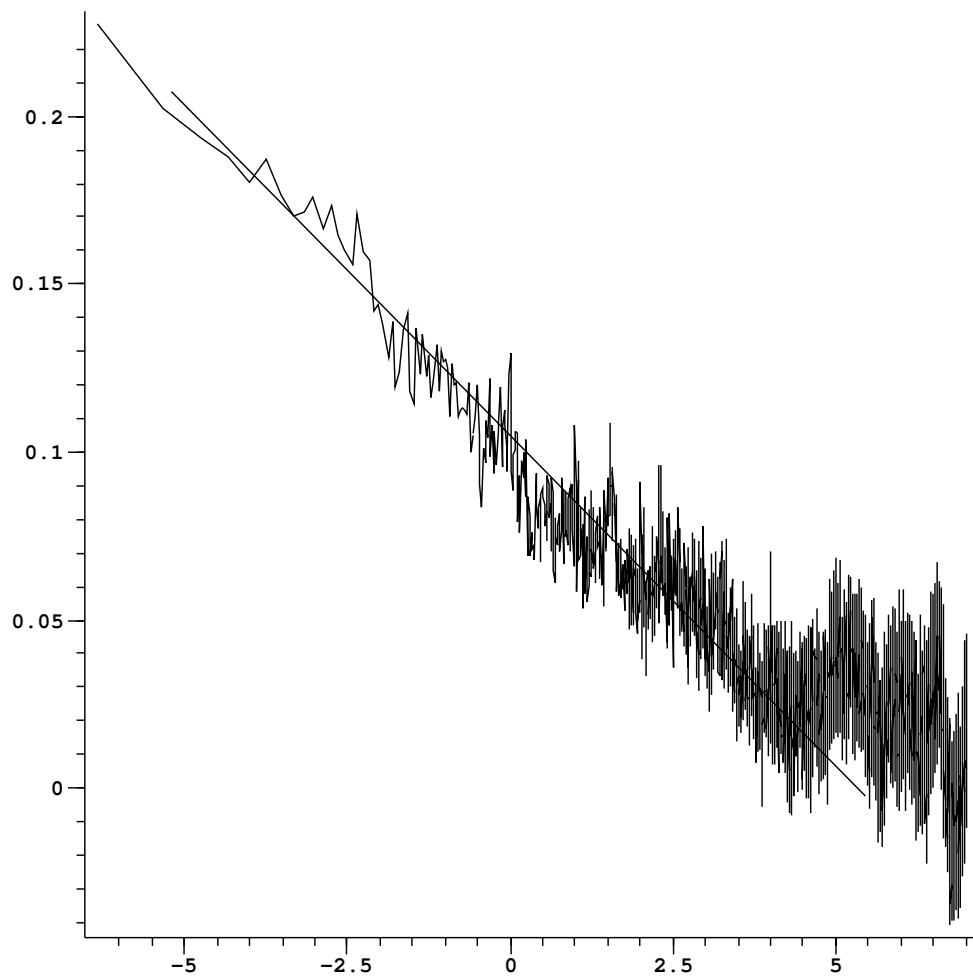




# Parameter Estimation

*(Bacry, Kozhemyak, Muzy 2004)*

S&P 500 intraday data (5mn ticks, 1996-1998)

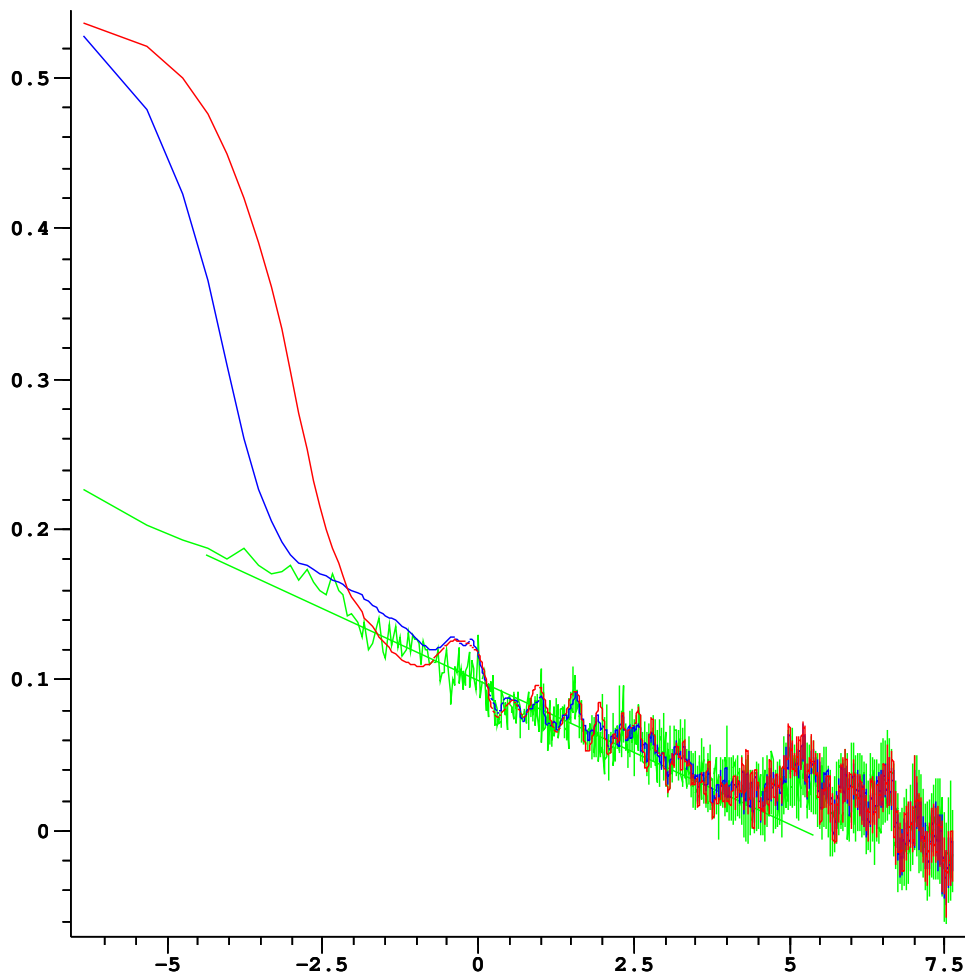


$\lambda^2 \simeq 0.03$  ,  $L \simeq 6$  months

# Parameter Estimation (changing $\Delta$ )

*(Bacry, Kozhemyak, Muzy 2004)*

S&P 500 intraday data (5mn ticks, 1996-1998)



— : 5mn returns  
— : 30mn returns  
— : 1h returns

# Other applications

(Bacry, Kozhemyak, Muzy 2004)

$$\delta X_{\Delta}[n] \simeq \epsilon[n] e^{\lambda \Omega_{\Delta}[n]/2}$$

- Variance prediction/estimation (Wiener filtering)

- Prediction of  $\Omega[n]$

$$\hat{\Omega}_{\Delta}[n] = h_2 * \ln(|\delta X_{\Delta}|)[n] \quad (h_2 \text{ causal})$$

→ MLE of  $e^{\lambda \Omega_{\Delta}[n]}$

- Value at Risk prediction/estimation

- estimation of  $v(p)$  such that

$$\text{Prob}\{|\delta X_{\Delta}[n]| > v(p) \mid \delta X_{\Delta}[n-k], k > 0\} = p$$

→ Edgeworth expansion